## AMENDMENTS TO THE CLAIMS

1. (Currently Amended) A method of detecting medical events in a medical instrument, comprising:

collecting ~~a plurality of~~ at least one training case[s] in the medical instrument, wherein ~~each~~ the training case has an input state indicative of at least a portion of a first biomedical signal of a particular patient and a corresponding output value indicative of a medical ~~condition~~ event of the particular patient;

~~generating~~ reconfiguring a neural network stored in the medical instrument based on the ~~plurality of~~ at least one training case[s] of the particular patient;

receiving a second biomedical signal of the particular patient in the medical instrument;

applying the second biomedical signal to the ~~generated~~ reconfigured neural network to generate an output of the neural network; ~~and~~

identifying a ~~condition~~ medical event of the particular patient based the output of the neural network; and

outputting data indicative of the identified medical event of the particular patient.

2. (Previously Presented) The method of claim 1, wherein collecting the plurality of training cases further comprises:

selecting a plurality of time epochs from a record of instrument feature values; and

indicating an output value for each selected time epoch.

3. (Previously Presented) The method of claim 2, wherein collecting the plurality of training cases further comprises:

selecting a configuration of instrument features; and wherein constructing the neural network based on the training cases comprises:

defining the neural network topology based on the input values and output values of the plurality of training cases; and

determining a kernel width value.

4. (Previously Presented) The method of claim 3, wherein training the neural network includes determining an optimal kernel width value by minimizing prediction error of the neural network.

5. (Previously Presented) The method of claim 4, wherein training the neural network further comprises:

determining an optimal input feature kernel width value for each input feature based on the determined optimal kernel width value.

6. (Original) The method of claim 3, wherein the neural network is a probabilistic neural network.

7. (Original) The method of claim 3, wherein the neural network is a generalized regression neural network.

8. (Original) The method of claim 3, wherein determining the kernel width value is based on a population statistic of the plurality of training cases.

9. (Original) The method of claim 8, wherein determining the kernel width value is based at least in part on the mathematical term of the number of training cases raised to an exponent power of about negative one-fifth.

10. (Original) The method of claim 9, wherein determining the kernel width value is based on the population distribution of the plurality of training cases.

11. (Original) The method of claim 10, wherein the population distribution of the plurality of training cases is approximately Normal.

12. (Previously Presented) The method of claim 3, further comprising normalizing the input values of the plurality of training cases based on the standard deviation for each input feature, and wherein determining the kernel width value comprises defining the kernel width value to be a number in the range 0.1 to 1.0.

13. (Original) The method of claim 3, wherein collection of the plurality of training cases further comprises:

normalizing the input values of the plurality of training cases based on the standard deviation for each input feature.

14. (Previously Presented) The method of Claim 1, wherein generating the neural network comprises a method for compressing the neural network, and wherein compressing the neural comprises:

determining a plurality of partitions based on the pattern layer nodes of the neural network wherein each partition comprises a plurality of groups of pattern layer nodes;

selecting one of the plurality of partitions based on a partition metric; and

for each group of pattern layer nodes within the selected partition:

replacing the group of pattern layer nodes with a compressed pattern layer node; and

adjusting the link weights between the compressed pattern layer node and any summation layer nodes to reflect the number of replaced pattern layer nodes.

15. (Previously Presented) The method of claim 14, wherein the partition metric comprises determining a BIC value for each partition.

16. (Previously Presented) The method of claim 15, wherein the partition metric comprises selecting the maximum BIC value.

17. (Original) The method of claim 14, wherein the partition metric comprises determining an error value for each partition.

18. (Original) The method of claim 14, wherein the partition metric comprises determining a compression ratio for each partition.

19. (Original) The method of claim 14, wherein the partition metric comprises determining a Minimum Description Length for each partition.

20. (Previously Presented) The method of claim 14, wherein the partition metric comprises determining a BIC value, an error value, and a compression ratio value for each partition.

21. (Original) The method of claim 20, wherein the K-means clustering method is applied to determine a plurality of partitions.

22. (Original) The method of claim 20, wherein the hierarchical clustering method is used to determine the plurality of partitions.

23. (Original) The method of claim 22, wherein the step of determining a plurality of partitions comprises applying the hierarchical clustering method to create partitions containing between about 1 and about 20 groups.

24. (Original) The method of claim 14, wherein selecting one of the determined plurality of partitions based on a partition metric comprises:

determining, for each partition within the determined plurality of partitions, a centroid value for each group of pattern layer nodes within that partition.

25. (Original) The method of claim 24, wherein selecting one of the determined plurality of partitions based on a partition metric further comprises:

determining, for each partition within the determined plurality of partitions, a covariance value for each group of pattern layer nodes within that partition.

26. – 30. (Canceled).

31. (Currently Amended) A method of incrementally updating a neural network based on correcting a prediction error, comprising:

applying the neural network in an electronic device to generate a first output value indicative of a classification of a first input state;

detecting a first prediction error in the first output value;

creating a first training case based on the first input state wherein the first training case corrects the first prediction error;

reconfiguring the neural network to correctly classify the first training case without altering the weights in retraining the neural network wherein reconfiguring the detection module further comprises adding a first pattern layer node to the neural network based on the first training case; ~~and~~

applying the neural network to generate a second output value from the electronic device indicative of a classification of a second input state; and

outputting data indicative of the second output value.

32. (Canceled).

33. (Previously Presented) The method of claim 31 wherein the neural network is initially incapable of correctly classifying a first input state.

34. (Canceled).

35. (Canceled).

36. (Canceled).

37. (Previously Presented) The method of claim 31, wherein reconfiguring the network comprises:

applying a detection module to classify ~~a~~ the first input state into a first event class;

determining that the detection module incorrectly classified the first input state into the first event class;

creating ~~a~~ the first training case by associating the first input state with a second event class; and

reconfiguring the detection module in real-time based on the first training case.

38. (Currently Amended) The method of claim 31, wherein ~~wherein~~ the second output from the electronic device comprises at least one of a display or a sound.

39. (Currently Amended) The method of claim 31 , wherein ~~wherein~~ the first and second input states are indicative of a biomedical signal of at least one patient and wherein the first and second output values are indicative of a medical ~~condition~~ event.

40. – 81. (Canceled).

82. (Currently Amended) A method of detecting medical ~~conditions~~ events in a patient, the method comprising:

receiving a biomedical signal of a particular patient;

identifying a portion of the signal that is indicative of a medical ~~condition~~ event of the particular patient based on user input; ~~and~~

~~generating~~ reconfiguring a predictive model stored in a memory of an electronic device for identifying a subsequent medical ~~condition~~ event of the particular patient based on an additional biomedical signal of the patient; and

storing the reconfigured predictive model in the memory of the electronic device.

83. (Previously Presented) The method Claim 82, wherein the biomedical signal comprises an electro-encephalogram.

84. (Currently Amended) The method Claim 82, wherein the medical ~~condition~~ event of the patient comprises a seizure.

85. (Currently Amended) The method of Claim 82, wherein ~~generating~~ reconfiguring the predictive model comprises ~~training~~ reconfiguring a neural network.

86. (Previously Presented) The method of Claim 82, wherein identifying the portion of the signal comprises:

at least partially displaying the signal; and

displaying at least one user control for selecting the identified portion of the signal.

87. (Currently Amended) The method of Claim 82, wherein identifying the portion of the signal comprises identifying an instrument feature of the signal.

88. (Currently Amended) The method of Claim 82, further comprising:

applying a second biomedical signal of the patient to the generated model to generate an output of the model; and

identifying the medical ~~condition~~ event of the patient based on the output of the model.

89. (Currently Amended) A system for detecting medical events from a record of instrument feature values, comprising:

a memory configured to store a neural network and ~~a plurality of~~ at least one training case[s], wherein ~~each~~ the training case has an input state indicative of at least a portion of a first biomedical signal of a particular patient and a corresponding output value indicative of a medical ~~condition~~ event of the particular patient; and

a processor configured to:

~~generate~~ reconfigure ~~a~~ the stored neural network based on the ~~plurality of~~ at least one training case[s] of the particular patient;

receive a second biomedical signal of the particular patient;

apply the second biomedical signal to the ~~generated~~ reconfigured neural network to generate an output of the neural network; and

identify a ~~condition~~ medical event of the particular patient based the output of the neural network; and

an output device configured to output data indicative of the identified medical ~~condition~~ event of the particular patient.

90. (Previously Presented) The system of Claim 89, wherein the biomedical signal comprises an electro-encephalogram.

91. (Currently Amended) The system of Claim 89, wherein the medical ~~condition~~ event of the patient comprises a seizure.

92. (Previously Presented) The system of Claim 89, wherein the output device comprises a display.

93. (Currently Amended) A system for detecting medical events from a record of instrument feature values, comprising:

means for storing a neural network and ~~a plurality of~~ at least one training case[s], wherein ~~each~~ the training case has an input state indicative of at least a portion of a first biomedical signal of a particular patient and a corresponding output value indicative of a medical ~~condition~~ event of the particular patient; ~~and~~

means for processing, said processing means configured to:

~~generate~~ reconfigure ~~a~~ the stored neural network based on ~~the plurality of~~ at least one training cases of a particular patient;

receive a second biomedical signal of the particular patient;

apply the second biomedical signal to the generated neural network to generate an output of the neural network; and

identify a ~~condition~~ medical event of the particular patient based the output of the neural network; and

means for outputting data indicative of the identified medical ~~condition~~ event of the particular patient.

94. (Previously Presented) The system of Claim 93, wherein the biomedical signal comprises an electro-encephalogram.

95. (Currently Amended) The system of Claim 93, wherein the medical ~~condition~~ event of the patient comprises a seizure.

96. (Previously Presented) The system of Claim 93, wherein the outputting means comprises a display.

97. (Previously Presented) The method of Claim 1, wherein the biomedical signal comprises an electro-encephalogram.

98. (Currently Amended) The method of Claim 1, wherein the medical ~~condition~~ event of the patient comprises a seizure.

99. (Currently Amended) A computer readable medium comprising instructions that when executed cause a processor to perform a method of detecting medical events, the method comprising:

collecting ~~a plurality of~~ at least one training case[s] in ~~the~~ a medical instrument, wherein ~~each~~ the training case has an input state indicative of at least a portion of a first biomedical signal of a particular patient and a corresponding output value indicative of a medical ~~condition~~ event of the particular patient;

~~generating~~ reconfiguring a neural network stored in the medical instrument based on the ~~plurality of~~ at least one training case[s] of the particular patient;

receiving a second biomedical signal of the particular patient in the medical instrument;

applying the second biomedical signal to the generated neural network to generate an output of the neural network; ~~and~~

identifying a ~~condition~~ medical event of the particular patient based the output of the neural network; and

outputting data indicative of the identified medical event of the particular patient.

100. (New) The method of Claim 1, wherein reconfiguring the neural network comprises adding a first pattern layer node to the neural network based on the at least one training case.

101. (New) The method of Claim 82, wherein reconfiguring the predictive model comprises adding a first pattern layer node to a neural network based on the additional biomedical signal of the patient.

102. (New) The system of Claim 89, wherein the processor is configured to reconfigure the neural network by adding a first pattern layer node to the neural network based on the at least one training case.

103. (New) The system of Claim 93, wherein the processing means is configured to reconfigure the neural network by adding a first pattern layer node to the neural network based on the at least one training case.

104. (New) The computer readable medium of Claim 99, wherein reconfiguring the neural network comprises adding a first pattern layer node to the neural network based on the at least one training case.